Discrete Probability Distributions. R

Administrator

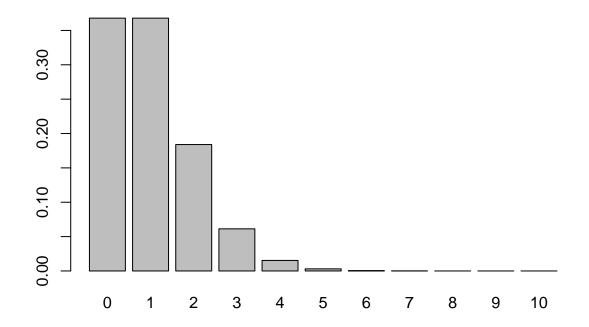
Tue Mar 22 14:51:05 2016

```
# Discrete Probability distributions and generating functions
# 18 March 2016
# NJG

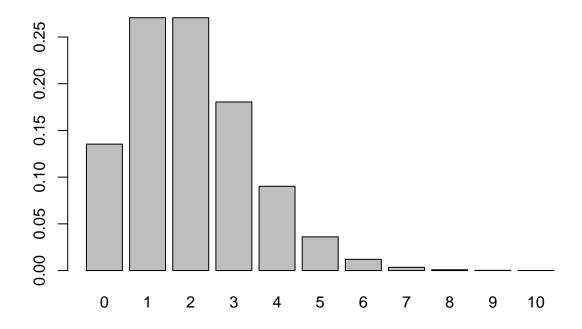
# Poisson distribution
# Discrete X >= 0
# Random events with a constant rate lambda
# (observations per time or per unit area)
# Parameter lambda > 0

# "d" function generates probability density

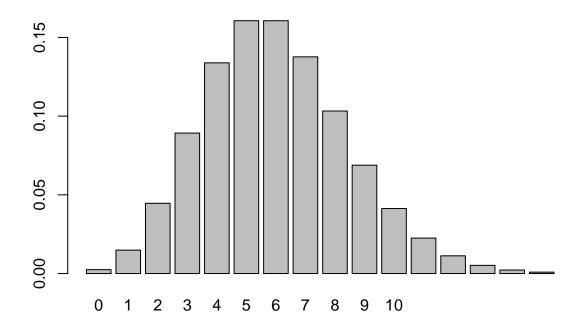
MyVec <- dpois(x=seq(0,10),lambda=1)
names(MyVec) <- seq(0,10)
barplot(height=MyVec)</pre>
```



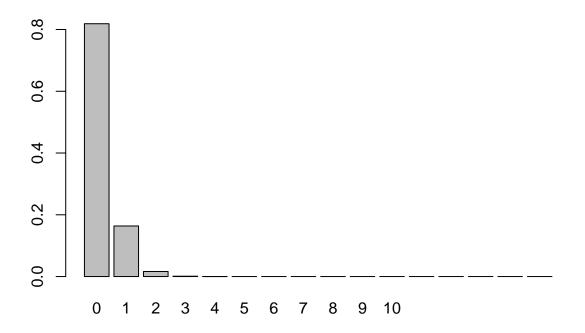
```
MyVec <- dpois(x=seq(0,10),lambda=2)
names(MyVec) <- seq(0,10)
barplot(height=MyVec)</pre>
```



```
MyVec <- dpois(x=seq(0,15),lambda=6)
names(MyVec) <- seq(0,10)
barplot(height=MyVec)</pre>
```



```
MyVec <- dpois(x=seq(0,15),lambda=0.2)
names(MyVec) <- seq(0,10)
barplot(height=MyVec)</pre>
```



```
sum(MyVec) # sum of density function = 1.0 (total area under curve)

## [1] 1

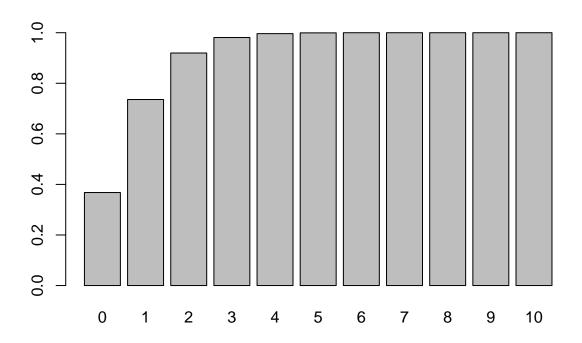
# for a Poisson distribution with lambda=2,
# what is the probability that a single draw will yield X=0?

dpois(x=0,lambda=2)

## [1] 0.1353353

# "p" function generates cumulative probability density; gives the
# "lower tail" cumulative area of the distribution

MyVec <- ppois(q=seq(0,10),lambda=1)
names(MyVec) <- seq(0,10)
barplot(height=MyVec)</pre>
```



```
# for a Poisson distribution with lambda=2,
# what is the probability that a single random draw will yield X <= 1?

ppois(q=1, lambda=2)

## [1] 0.4060058

# to generate the "upper tail", use lower.tail=FALSE
# important: lower tail P(X <= z) but upper tail P(X > z)

# Therefore for any z, P(lower.tail z) + P(upper.tail z) = 1.0

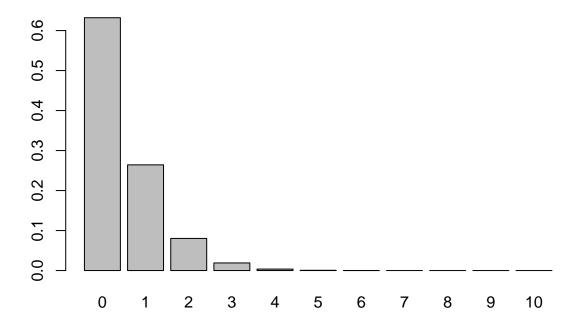
z <- 3

ppois(q=3,lambda=1) + ppois(q=3,lambda=1,lower.tail=FALSE)

## [1] 1

MyVec <- ppois(q=seq(0,10),lambda=1,lower.tail=FALSE)

names(MyVec) <- seq(0,10)
barplot(height=MyVec)</pre>
```



```
# for a Poisson distribution with lambda=2,
# what is the probability that a single random draw will yield X >= 4?

ppois(q=3,lambda=2,lower.tail=FALSE)
```

[1] 0.1428765

```
# Because the total area under the curve is 1.0,
# we can subtract the two tail probabilities from 1.0
# to obtain the probability of a value being in a certain interval.

# for a Poisson distribution with lambda = 4,
# what is the probability that a single random draw
# will yield a value X that is between 1 and 3?

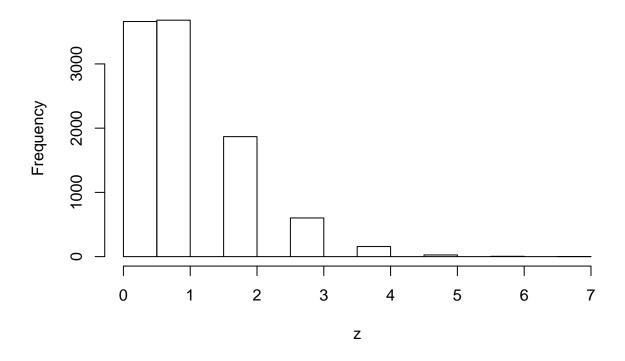
lowerTail <- ppois(q=0,lambda=4)
upperTail <- ppois(q=3,lambda=4, lower.tail=FALSE)

IntervalP <- 1.0 - lowerTail - upperTail
print(IntervalP)</pre>
```

[1] 0.4151545

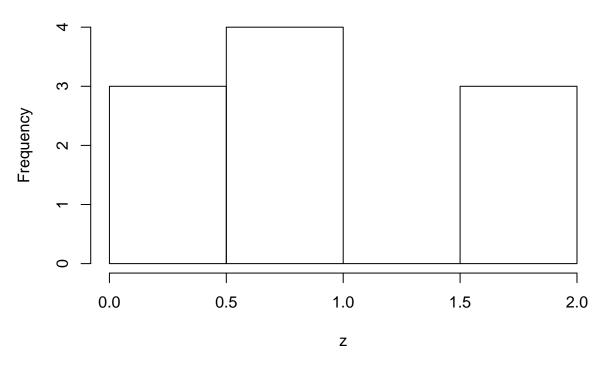
```
# Lets check this by calculating and summing
# the probabilities for the individual integer values
x1 <- dpois(x=1,lambda=4)</pre>
x2 <- dpois(x=2,lambda=4)</pre>
x3 <- dpois(x=3,lambda=4)
sum(x1,x2,x3)
## [1] 0.4151545
# the "q" quantile function is the "inverse" of the "p" function.
# We give a tail probability and it gives us a
# value X that would correspond to that.
# the quantile is right continuous
# qpois(p, lambda) is the smallest integer such that P(X \le z) >= p.
qpois(p=0.75,lambda=3)
## [1] 4
# The q function can be used to generate a 95% interval on the distribution:
qpois(p=c(0.025,0.975),lambda=10)
## [1] 4 17
qpois(p=0.975,lambda=10,lower.tail=TRUE)
## [1] 17
# We can also find the midpoint of probability in the distribution:
qpois(p=0.5,lambda=3)
## [1] 3
qpois(p=0.5,lambda=0.1)
## [1] 0
# Use rpois to generate a random set of values from a distribution:
z <- rpois(n=10000,lambda=1)</pre>
hist(z,breaks=10)
```

Histogram of z

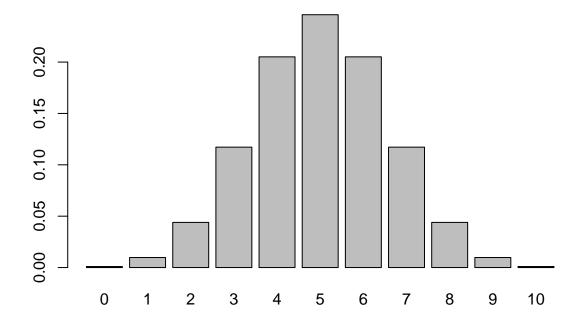


z <- rpois(n=10,lambda=1)
hist(z)</pre>

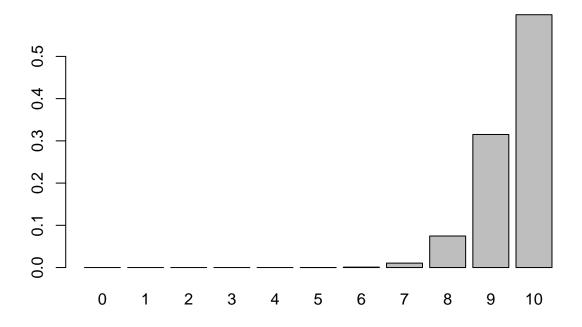
Histogram of z



```
# Make comparisons of sample values with predicted values
z <- rpois(n=10000,lambda=1) # very large sample</pre>
table(z)
                              # get frequencies
## z
           1
                      3
                                5
                                      6
                                           7
                2
## 3758 3593 1867 601
                                     8
                        136
                               36
                                           1
mean(z==5)
                              # get proportion of values = 5
## [1] 0.0036
dpois(x=5,lambda=1)
                              # get theoretical proportion
## [1] 0.003065662
# Binomial distribution
# p = probability of a dichotomous outcome
# size = number of trials
\# x = possible outcomes
# use "d" binom for density function
MyVec <- dbinom(x=seq(0,10),size=10,prob=0.5)</pre>
names(MyVec) <- seq(0,10)</pre>
barplot(height=MyVec)
```



```
MyVec <- dbinom(x=seq(0,10),size=10,prob=0.95)
names(MyVec) <- seq(0,10)
barplot(height=MyVec)</pre>
```



```
# use "p" binom for cumulative distribution
# what is probability of getting 5 heads out of 10 tosses?
dbinom(x=5,size=10,prob=0.5)

## [1] 0.2460938
# what is the probability of getting 5
# or fewer heads out of 10 tosses?
pbinom(q=5,size=10,prob=0.5)

## [1] 0.6230469
pbinom(q=4,size=9,prob=0.5)

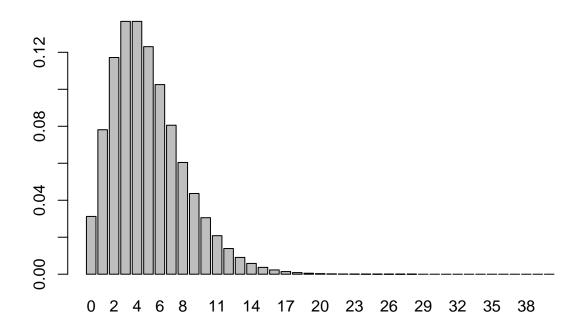
## [1] 0.5
# use "q" binom for quantiles
# what minimum number of heads will be found
# for 40% of 50 trials with p = 0.5?
qbinom(p=0.4,size=50,prob=0.5)
```

[1] 24

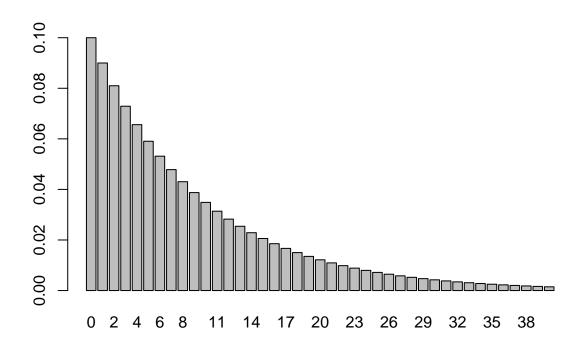
```
# what is a 95% confidence interval for 100 trials
# of a coin with p = 0.7 for heads?
qbinom(p=c(0.025,0.975),size=100,prob=0.7)
```

[1] 61 79

```
# negative binomial: number of failures (values of MyVec)
# in a series of (Bernouli) with p=probability of success
# before a target number of successes (= size)
# generates a discrete distribution that is more
# heterogeneous ("overdispersed") than Poisson
MyVec <- dnbinom(x=seq(0,40), size=5, prob=0.5)
names(MyVec) <- seq(0,40)
barplot(height=MyVec)</pre>
```

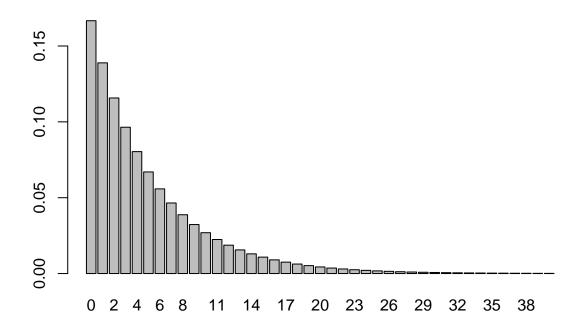


```
# geometric series is a special case where N= 1 success
# each bar is a constant fraction 1 - "prob" of the bar before it
MyVec <- dnbinom(x=seq(0,40), size=1, prob=0.1)
names(MyVec) <- seq(0,40)
barplot(height=MyVec)</pre>
```



```
# alternatively specify mean = mu of distribution and size,
# the dispersion parameter (small is more dispersed)

MyVec <- dnbinom(x=seq(0,40),size=1,mu=5)
names(MyVec) <- seq(0,40)
barplot(height=MyVec)</pre>
```



```
# also have the "pnbinom", "qnbinom" and "rnbinom" functions
# Probability of drawing a 3 or smaller from a negative binomial:
pnbinom(q=3,size=1,mu=5)

## [1] 0.5177469

## 5 percent lower value for a negative binomial
qnbinom(p=0.05,size=10,mu=5)

## [1] 1

# 95% confidence interval for a geometric series
qnbinom(p=c(0.025,0.975),prob=0.5,size=10)

## [1] 3 20

# random sample from a negative binomial

MyVec <- rnbinom(n=1000,size=1,mu=20)
quantile(MyVec,prob=c(0.025,0.975))

## 2.5% 97.5%
## 0.00 88.05</pre>
```

```
# compare to exact calculation
qnbinom(p=c(0.025,0.975),size=1,mu=20)
## [1] 0 75
# Sampling from a vector of elements (not necessarily numbers)
Species <- LETTERS
print(Species)
## [1] "A" "B" "C" "D" "E" "F" "G" "H" "I" "J" "K" "L" "M" "N" "O" "P" "Q"
## [18] "R" "S" "T" "U" "V" "W" "X" "Y" "Z"
# Sampling with or without replacement
# Sampling equiprobably or with specified (unequal) probabilities)
# Case 1: Sampling equiprobably with replacement
sample(Species, size=10, replace=TRUE)
## [1] "M" "V" "R" "H" "R" "K" "Y" "Y" "B" "I"
# Case 2: Sampling equiprobably without replacement
sample(Species, size=10, replace=FALSE)
## [1] "D" "O" "R" "V" "C" "Z" "G" "E" "P" "M"
# Case 3: Sampling with specified probabilities with replacement
# any set of non-negative numbers
# that is the same length as the vector)
MyProbs \leftarrow seq(1,26)
sample(Species, size=10,replace=TRUE,prob=MyProbs)
## [1] "N" "V" "M" "R" "S" "N" "T" "W" "Y" "D"
# Case 4: Sampling with specified probabilities, without replacement
sample(Species, size=10,replace=FALSE,prob=MyProbs)
## [1] "D" "S" "Y" "P" "R" "O" "Q" "E" "X" "Z"
```